

# Pedestrians: the new kings of smart cities

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# Outline

- 1 Motivation
- 2 Fundamental quantities
  - Discretization
  - 3D Voronoi
  - Indicators
- 3 Moving walkways



# Motivation



# Motivation

## A world of cities

- 2014: 54% of the world's population lives in cities Source: UN

## Share of walking trips in cities

- |                       |                        |
|-----------------------|------------------------|
| ● Bangalore 2011: 26% | ● Barcelona 2006: 38%  |
| ● Beijing, 2011: 21%  | ● Berlin, 2010: 29%    |
| ● Bogota, 2008: 15%   | ● Chicago, 2008: 19%   |
| ● Delhi, 2011: 21%    | ● Madrid, 2006: 36%    |
| ● London, 2011: 30%   | ● Singapore, 2011: 22% |
| ● New-York, 2010: 39% | ● Mumbai, 2011: 27%    |

Source: [LTA Academy, 2011]

# Research challenges



- Understand, describe and predict
- Design of facilities
- Management and control
- Information and guidance

# In this talk...

- 1 Characterization of fundamental quantities
- 2 A futuristic transportation system: a network of moving walkways

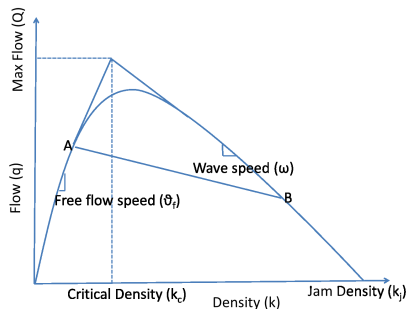


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# Fundamental quantities



For pedestrians

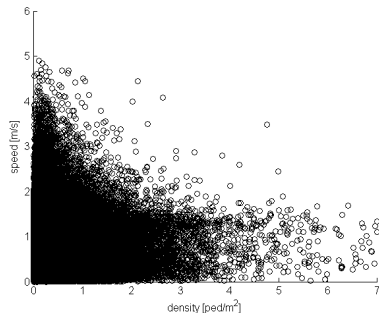
- Density  $k$  (ped/m<sup>2</sup>)
- Speed  $v$  (m/s)
- Flow  $q$  (ped/ms)



# Pedestrians $\neq$ vehicles

## Issues

- Scattered fundamental diagram
- Impact of spatial discretization

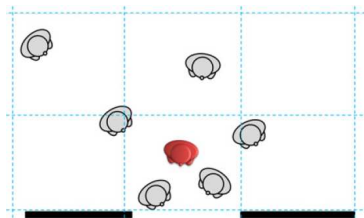


25603 trajectories, Lausanne train station, February 2013

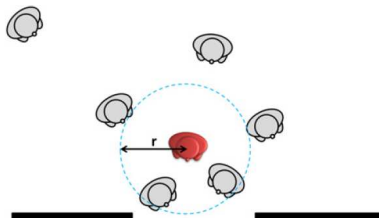
Source: [Nikolic et al., 2016]



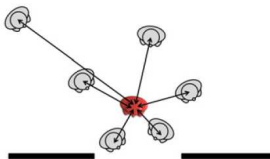
# Discretization methods



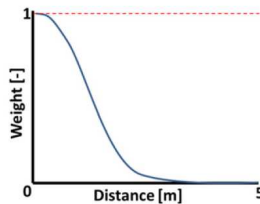
Grid-based (GB)



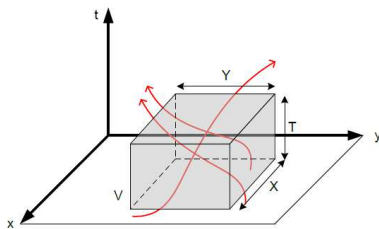
Range-based (RB)



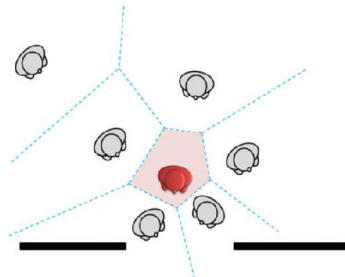
Exponentially Weighted (EW)



# Discretization methods



**Edie (XY-T)**



**Voronoi-based (VB)**

# Context

## Model

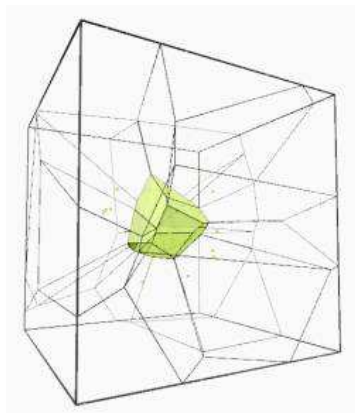
- Space-time representation:  $\Omega \subset \mathbb{R}^3$
- Units: meters and seconds
- $p = (x, y, t) \in \Omega$ : physical position  $(x, y)$  in space at a specific time  $t$
- Assumption:  $\Omega$  is convex (obstacle-free and bounded)

## Data: trajectories

- Continuous:  $\Gamma_i : \{p_i(t) | p_i(t) = (x_i(t), y_i(t), t)\}$
- Discrete (sample):  $\Gamma_i : \{p_{is} | p_{is} = (x_{is}, y_{is}, t_s)\}, t_s = [t_0, t_1, \dots, t_f]$



# 3D Voronoi diagram



## Definition

- For each point  $p \in \Omega$
- For each trajectory  $\Gamma_i$
- Define a distance  $D(p, \Gamma_i)$
- Associate  $p$  with the closest trajectory:  
 $\delta_{\Gamma}(p, \Gamma_i) =$

$$\begin{cases} 1, & D(p, \Gamma_i) \leq D(p, \Gamma_j), \forall j \neq i \\ 0, & \text{otherwise} \end{cases}$$

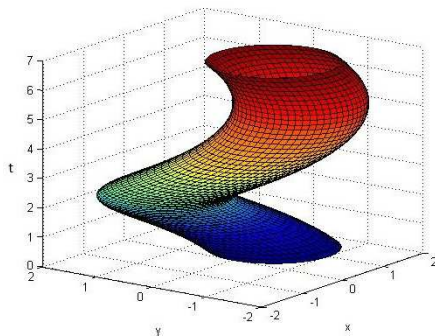
# 3D Voronoi diagram

## Distance

$$D(p, \Gamma_i) = \min_{p_i \in \Gamma_i} \{d(p, p_i)\},$$

- Various definitions of  $d(\cdot, \cdot)$  are possible. [Nikolic and Bierlaire, 2016]
- Voronoi cell for trajectory  $i$ :

$$V_i = \{p \in \Omega | \delta_{\Gamma}(p, \Gamma_i) = 1\}$$



# Intersection with a plane

## Notation

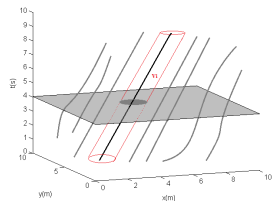
$\mathcal{P}_{(a,b,c),p_0}$ : plane through  $p_0$  with normal vector  $(a, b, c)$



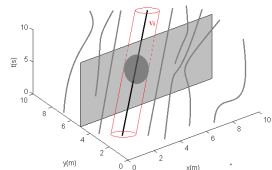
# Intersection with a plane

## Intersections

Intersection with  $\mathcal{P}_{(0,0,1),p_0}$



Intersection with  $\mathcal{P}_{(a,b,0),p_0}$





# Voronoi-based traffic indicators

Consider  $(x, y, t) \in \Omega$ , and  $i$  such that  $(x, y, t) \in V_i$ .

## Density indicator

$$k(x, y, t) = \frac{1}{|V_i \cap \mathcal{P}_{(0,0,1),(x,y,t)}|}$$

## Flow indicator

$$\vec{q}_{(a,b,0)}(x, y, t) = \frac{1}{|V_i \cap \mathcal{P}_{(a,b,0),(x,y,t)}|}$$

## Velocity indicator

$$\vec{v}_{(a,b,0)}(x, y, t) = \frac{\vec{q}_{(a,b,0)}(x, y, t)}{k(x, y, t)} = \frac{|V_i \cap \mathcal{P}_{(0,0,1),(x,y,t)}|}{|V_i \cap \mathcal{P}_{(a,b,0),(x,y,t)}|}$$

# Main findings

- Data driven discretization.
- Well defined and flexible.
- Robust to noise in the data.
- Robust to sampling of trajectories.
- Details in [Nikolic and Bierlaire, 2016].



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# Cars: kings of our cities



## Surface used by streets and parkings

- Houston, TX: 64.7%
- Little Rock, AR: 61.2%
- Milwaukee, WI: 54.1%
- Washington, DC: 44.4%

Source: [Gardner, 2011]

# What about a “post car” world?

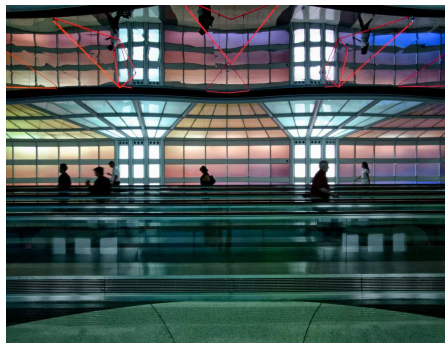
- Cars are banned from cities.
- The surface of streets is claimed for pedestrians.
- Problem: speed.
- Possible solution: moving walkways



# Paris, 1900



# Moving walkways



## Sustainable

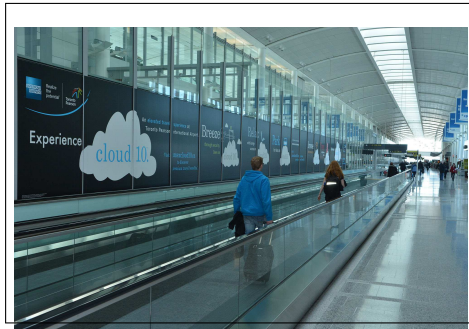
- Electric
- No local emission
- Energy efficient

## Functional

- Continuous flow
- Speed: accelerated moving walkways



# Toronto Airport, today





# Costs

System	Capital cost [M EUR/km]	Typical costs [EUR/pax-km]	Operational cost [EUR/pax-km]
Bus	0.1- 6.7	1500	0.09-0.95
Light rail	8.5-83.5	2800	0.07-0.28
PRT	6.7-25.4	3500	0.07-0.28
AMW	34.8-54.4	7300	0.08-0.42

- ✗ High capital costs
- ✗ High typical costs
- ✓ Competitive operational costs



# Efficiency

System	Average speed [km/h]	Capacity [pax/h]	Corridor width [m]
Bus	15-20	1,000-4,500	3.0-4.2
Light rail	15-45	1,000-30,000	2.5-3.2
PRT	20-25	1,800-7,200	2.5-3.2
AMW	5-12	4,500-7,500	1.2-2.3

- ✓ Competitive speed
- ✓ High capacity
- ✓ Low space usage



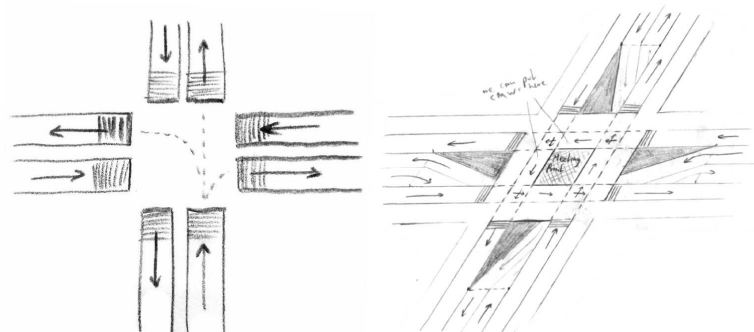
# Energy

System	Energy use [MJ/pax-km]	Noise level [dB(A)]
Bus	0.30-1.56	70-84
Light rail	0.70-2.50	60-74
PRT	0.55	35-65
AMW	0.11	54

- ✓ Low energy consumption
- ✓ Low noise level



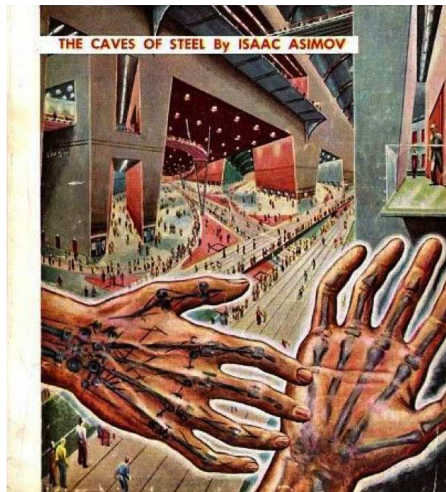
# Network design



## Case study; Geneva

- Two objectives: mobility and costs.
- Good trade off with 44 AMWs.
- Details in [Scarinci et al., 2014] and [Scarinci et al., 2016].

# Pedestrians: new kings of smart cities?



## Data

Pedestrian trajectories

## Technology

Accelerated moving walkways

## Models

Specification, validation, prediction

## Urban Systems

Integration

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